Storage and Indexing

Introduction to Databases

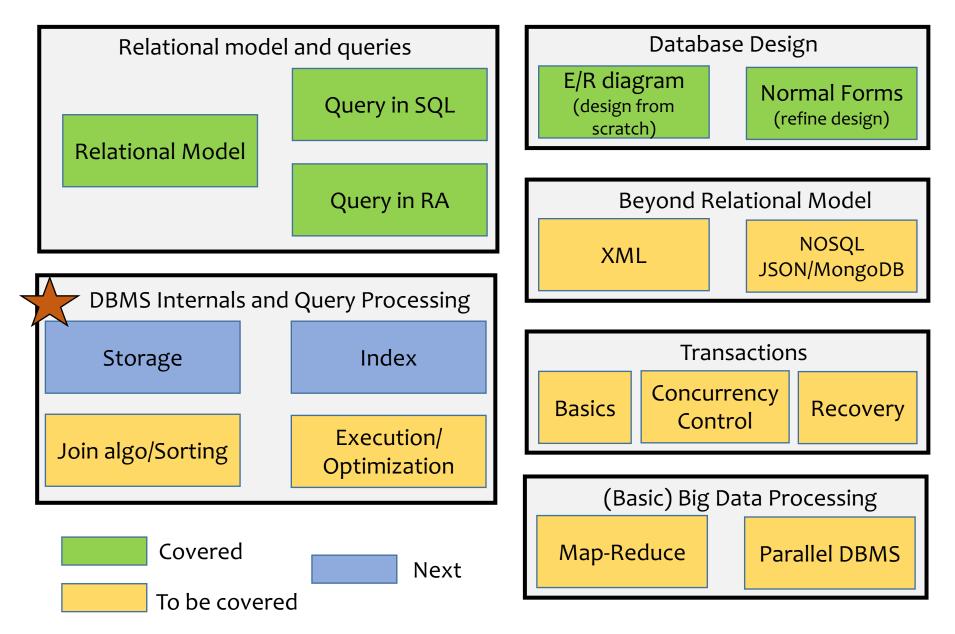
CompSci 316 Fall 2020



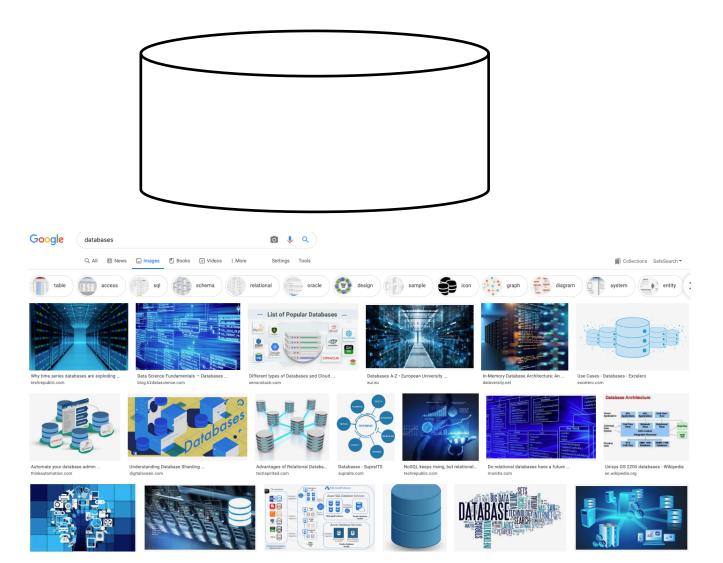
Announcements (Thu. Oct 1)

- Keep working on your project!
 - MS-2 due in two weeks (10/15)
 - Need to submit a basic working version of your website (all functionalities not needed, but interactions from/to UI and databases should be there)+ other things
- HW-5/Gradiance-3 to be released today
 - Due in a week 10/8 (Thu)

Where are we now?



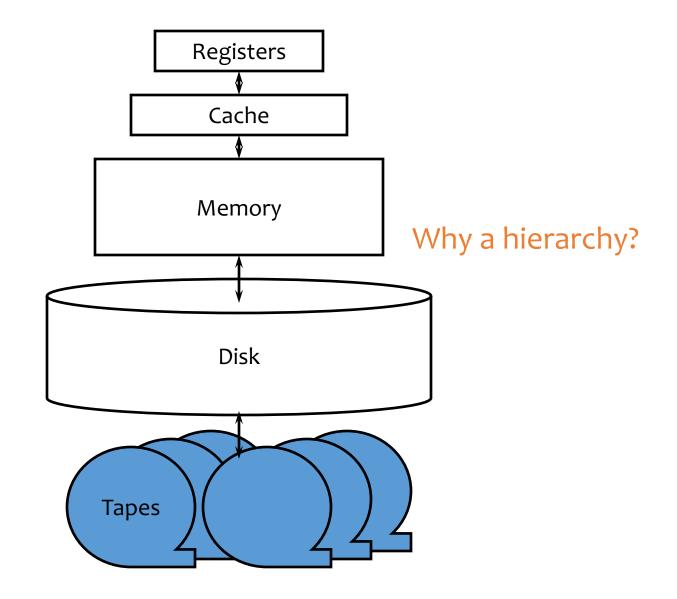
Why do we draw databases like this?



Outline

- It's all about disks!
 - That's why we always draw databases as
 - And why the single most important metric in database processing is (oftentimes) the number of disk I/O's performed

Storage hierarchy



How far away is data?

Location	<u>Cycles</u>	Location	<u>Time</u>
Registers	1	My head	1 min.
On-chip cache	2	This room	2 min.
On-board cache	10	Duke campus	10 min.
Memory	100	Washington D.C.	1.5 hr.
Disk	10 ⁶	Pluto	2 yr.
Таре	10 ⁹	Andromeda	2000 yr.

(Source: AlphaSort paper, 1995) The gap has been widening!

I/O dominates—design your algorithms to reduce I/O!

Latency Numbers Just FYI – Take a look yourself! Every Programmer Should Know

Latency Comparison Numbers

L1 cache reference	0.5	ns ns			
Branch mispredict	5	ns			
L2 cache reference	7	ns			14x L1 cache
Mutex lock/unlock	25	ns			
Main memory reference	100	ns			20x L2 cache, 200x L1 cache
Compress 1K bytes with Zippy	3,000	ns	3 us		
Send 1K bytes over 1 Gbps network	10,000	ns	10 us		
Read 4K randomly from SSD*	150,000	ns	150 us		~1GB/sec SSD
Read 1 MB sequentially from memory	250,000	ns	250 us		
Round trip within same datacenter	500,000	ns	500 us		
Read 1 MB sequentially from SSD*	1,000,000	ns	1,000 us	1 ms	~1GB/sec SSD, 4X memory
Disk seek	10,000,000	ns	10,000 us	10 ms	20x datacenter roundtrip
Read 1 MB sequentially from disk	20,000,000	ns	20,000 us	20 ms	80x memory, 20X SSD
Send packet CA->Netherlands->CA	150,000,000	ns	150,000 us	150 ms	

Notes

```
1 ns = 10<sup>-9</sup> seconds
1 us = 10<sup>-6</sup> seconds = 1,000 ns
1 ms = 10<sup>-3</sup> seconds = 1,000 us = 1,000,000 ns
```

Credit

```
_____
```

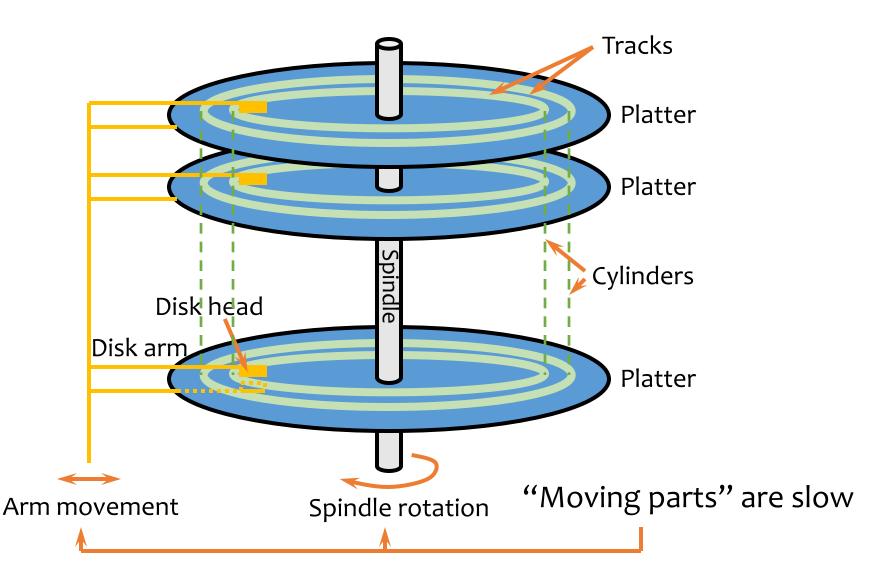
```
By Jeff Dean: http://research.google.com/people/jeff/
Originally by Peter Norvig: http://norvig.com/21-days.html#answers
```

A typical hard drive



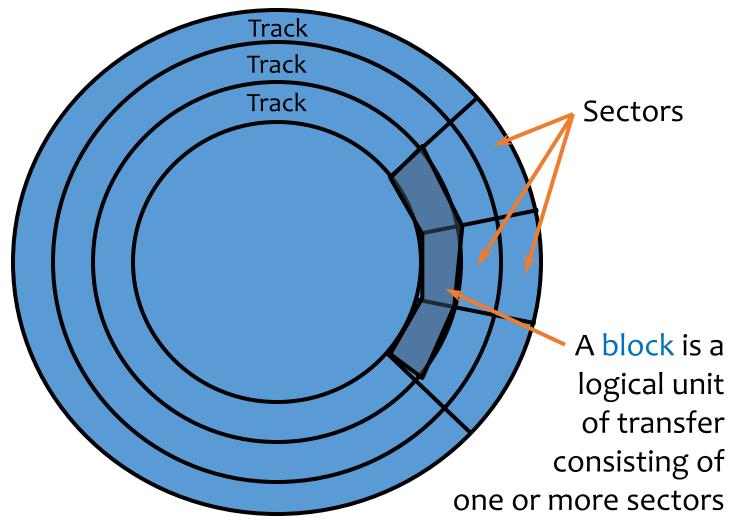
http://upload.wikimedia.org/wikipedia/commons/f/f8/Laptop-hard-drive-exposed.jpg

A typical hard drive



Top view

"Zoning": more sectors/data on outer tracks



Disk access time

Sum of:

- Seek time: time for disk heads to move to the correct cylinder
- Rotational delay: time for the desired block to rotate under the disk head
- Transfer time: time to read/write data in the block (= time for disk to rotate over the block)

Sequential vs. Random disk access

Seek time + rotational delay + transfer time

- Average seek time
 - Sequential: 0
 - Random: "Typical" value: 5 ms
- Average rotational delay
 - Sequential: o
 - Random: "Typical" value: 4.2 ms (7200 RPM)
- Transfer time
 - Thee same for sequential and random
- Sequential is an order of magnitude faster!

Important consequences

- It's all about reducing I/O's!
- Cache blocks from stable storage in memory
 - DBMS maintains a memory buffer pool of blocks
 - Reads/writes operate on these memory blocks
 - Dirty (updated) memory blocks are "flushed" back to stable storage

Picture on board that we will use again and again!

Performance tricks

- Disk layout strategy
 - Keep related things (what are they?) close together: same sector/block → same track → same cylinder → adjacent cylinder
- Prefetching
 - While processing the current block in memory, fetch the next block from disk (overlap I/O with processing)
- Parallel I/O
 - More disk heads working at the same time
- Disk scheduling algorithm
 - Example: "elevator" algorithm
- Track buffer
 - Read/write one entire track at a time

Not covered in detail

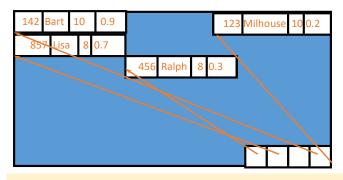
Data layout on disk

How each component is stored in the parent Table \rightarrow Pages/Blocks \rightarrow Records/Tuples/Rows \rightarrow Attributes

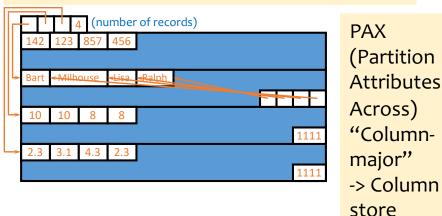
Examples:

0	- T				24	2	.8	36
	142Bart (padded with space)					10	0.9	
	Fixe	d-len	gth fie	elds				
0	9 4	. 8	3	16				
	142	10	0.9		Bart\0	W	eird kid!\0	
C) 2	1 8	8	1	6 18	22		32
	142	10	0.9)	Ba	rt	Weird kid!	
22 32								
Variable-length fields (delimiter or offset ar								

rray)



N-ary storage model/NSM "Row-major", directory at the end Reorganization needed after updates



Take-away

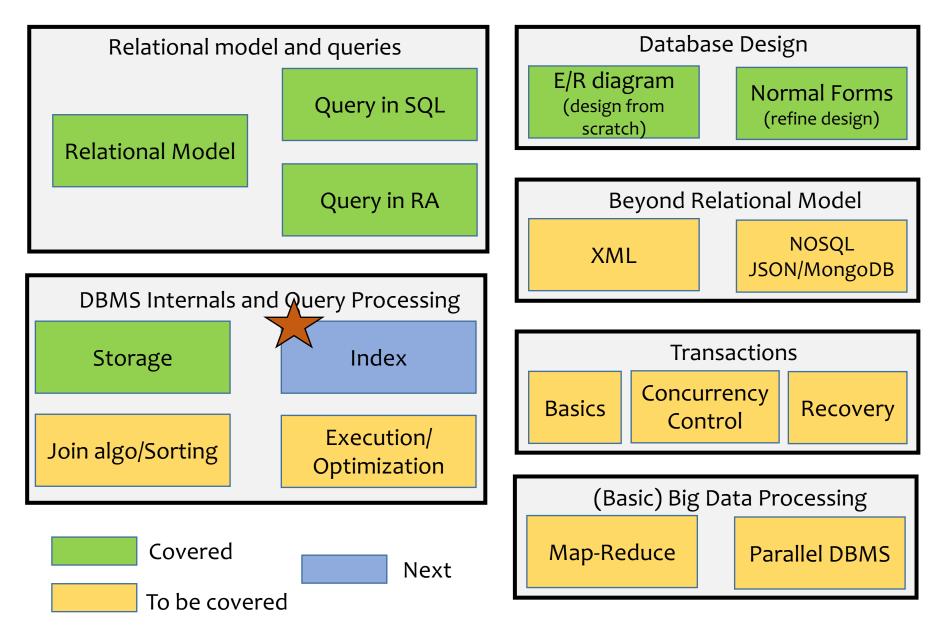
- Storage hierarchy
 - Why I/O's dominate the cost of database operations
- Disk
 - Steps in completing a disk access
 - Sequential versus random accesses
- Disk is slower than Main memory = Buffer Pool
 - Minimize the number of transfers to/from Disk
 - Our unit of cost!
 - All computation cost ignored by default

Index

Announcements (Tue. Oct 6)

- HW-5 + Gradiance-3 (Constraints/Triggers)
 - Due this Friday 10/9
- Keep working on your project!
 - MS-2 due next week (10/15)
 - Need to submit a basic working version of your website (all functionalities not needed, but interactions from/to UI and databases should be there) + other things
- If you would like to meet me one-one, please email Yesenia and me ASAP
 - By tomorrow (Wed 10/7)

Where are we now?



Recall the Disk-Main Memory diagram!

Topics

- Index
- Dense vs. Sparse
- Clustered vs. unclustered
- Primary vs. secondary
- Tree-based vs. Hash-index

Related

What are indexes for?

- Given a value, locate the record(s) with this value
 SELECT * FROM R WHERE A = value;
 SELECT * FROM R, S WHERE R.A = S.B;
- Find data by other search criteria, e.g.
 - Range search
 SELECT * FROM *R* WHERE *A* > *value*;
 - Keyword search

Focus of this lecture

database indexing

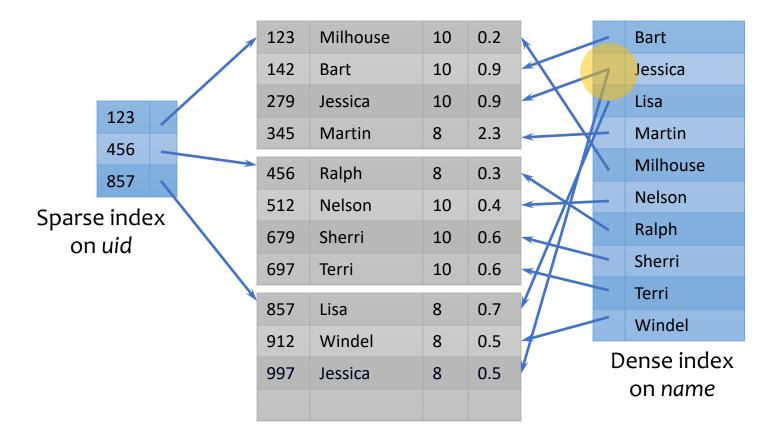


Dense and sparse indexes

When are these possible?

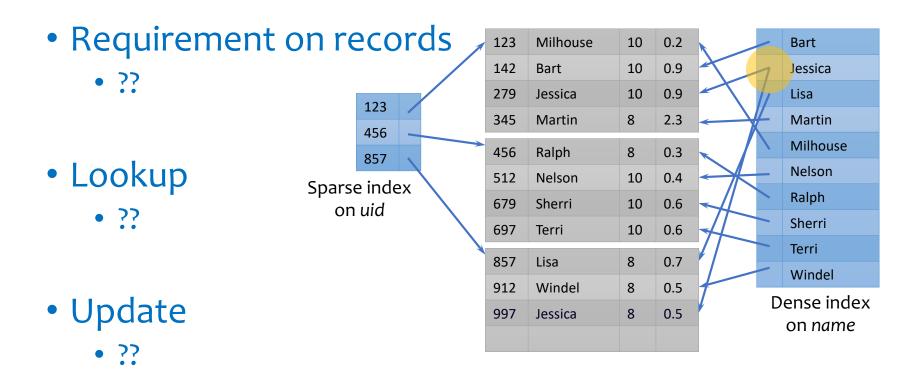
Comparison?

- Dense: one index entry for each search key value
 - One entry may "point" to multiple records (e.g., two users named Jessica)
- Sparse: one index entry for each block
 - Records must be clustered according to the search key



Dense versus sparse indexes

- Index size
 - ??



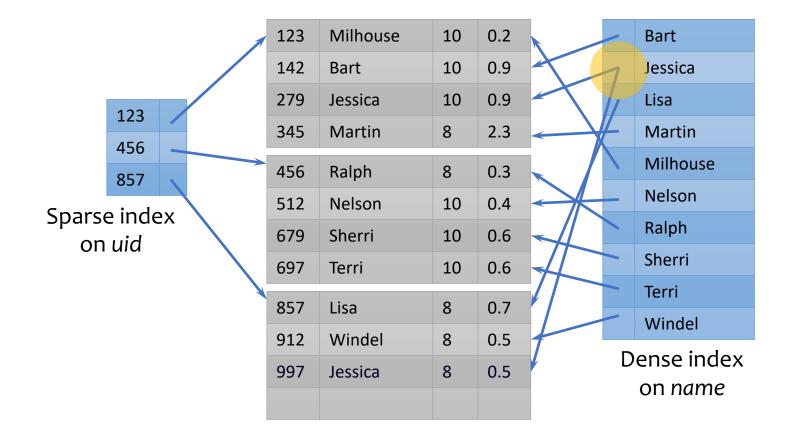
Dense versus sparse indexes

- Index size
 - Sparse index is smaller
- Requirement on records
 - Records must be clustered for sparse index
- Lookup
 - Sparse index is smaller and may fit in memory
 - Dense index can directly tell if a record exists
- Update
 - May be easier for sparse index (less movement for updates)

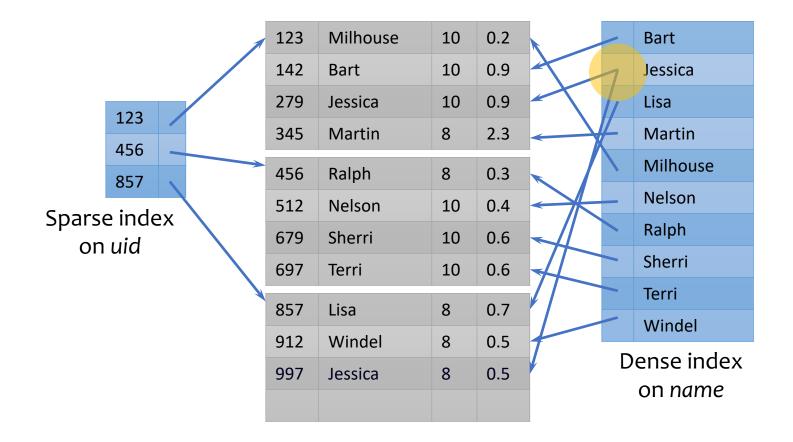
Primary and secondary indexes

- Primary index
 - Created for the primary key of a table
 - Records are usually clustered by the primary key
 - Can be sparse
- Secondary index
 - Usually dense
- SQL
 - PRIMARY KEY declaration automatically creates a primary index, UNIQUE key automatically creates a secondary index
 - Additional secondary index can be created on non-key attribute(s): CREATE INDEX UserPopIndex ON User(pop);

What if the index is too big as well?



What if the index is too big as well?

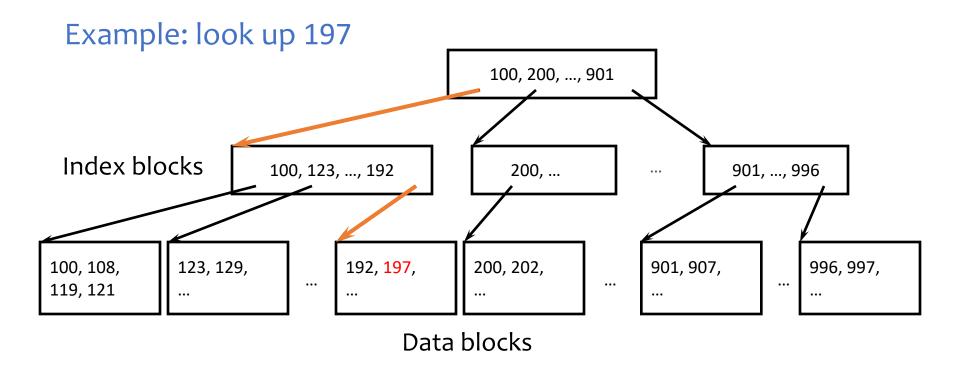


Put a another (sparse) index on top of that!

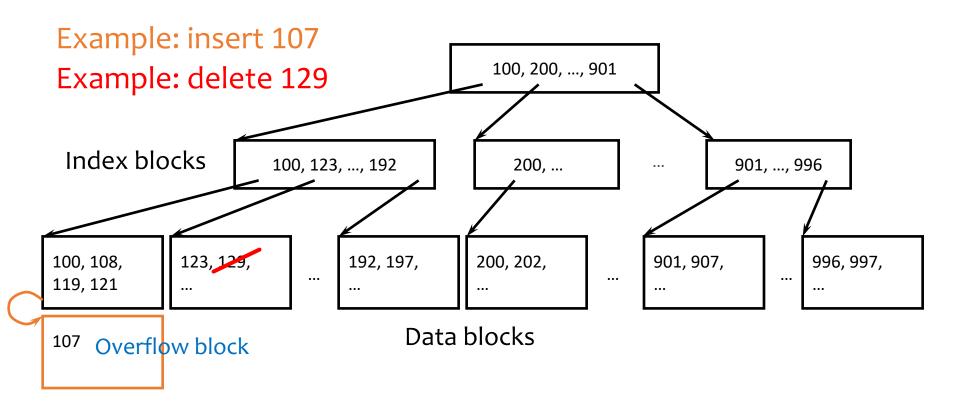
ISAM

- What if an index is still too big?
 - Put a another (sparse) index on top of that!

ISAM (Index Sequential Access Method), more or less

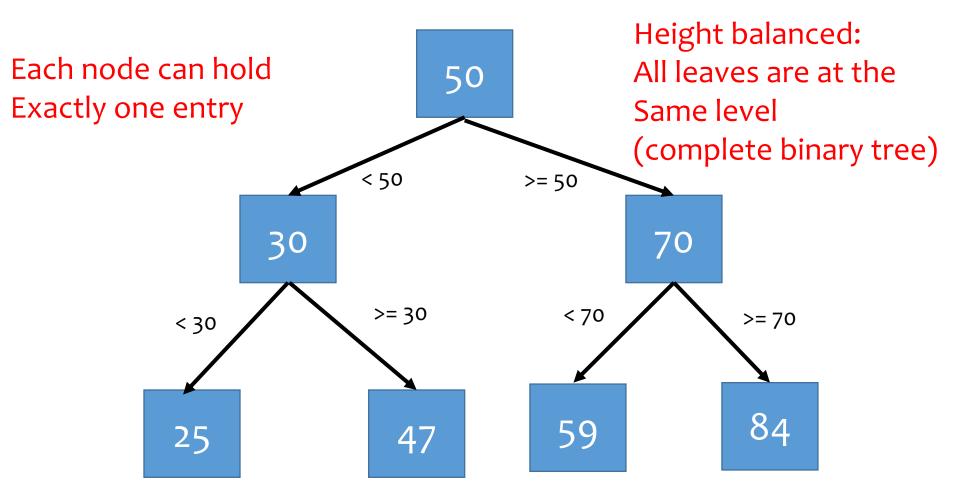


Updates with ISAM



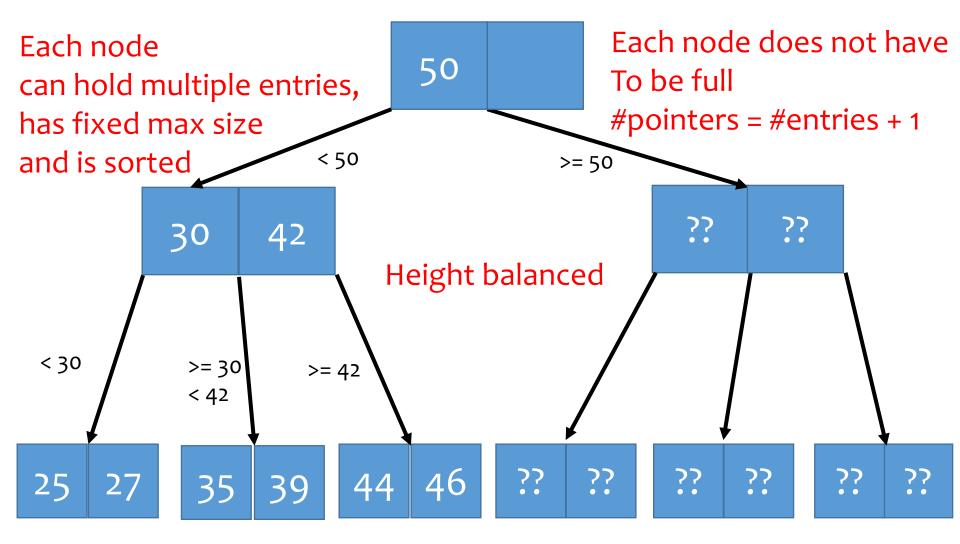
- Overflow chains and empty data blocks degrade performance
 - Worst case: most records go into one long chain, so lookups require scanning all data!

Binary Search Tree



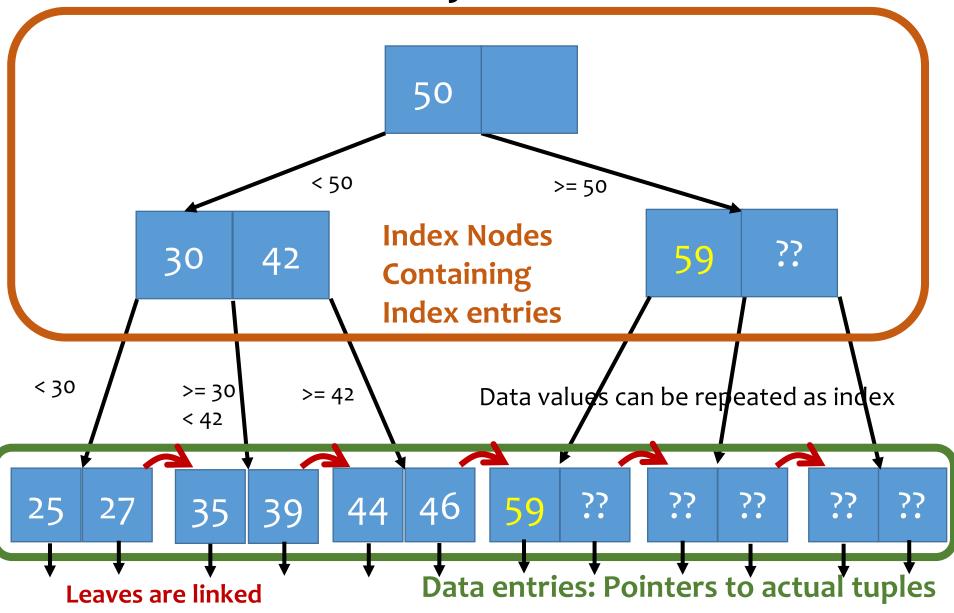
Leaves are sorted

B-tree: Generalizing Binary Search Trees



Leaves are sorted

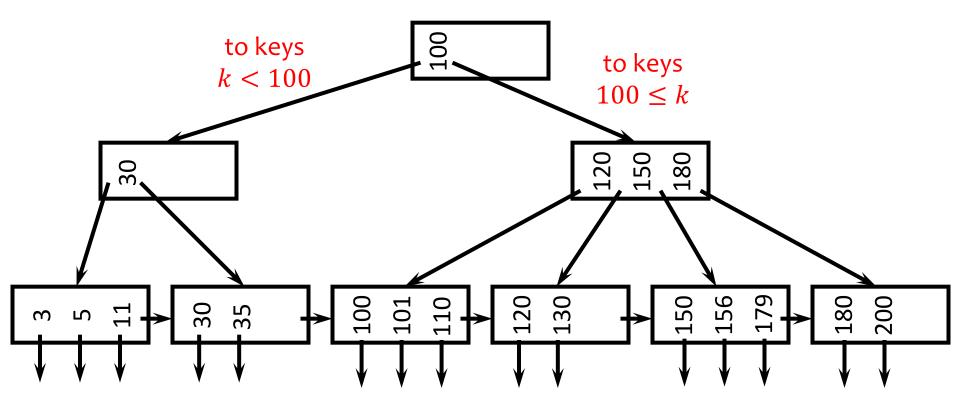
B⁺-tree: Data only at leaves



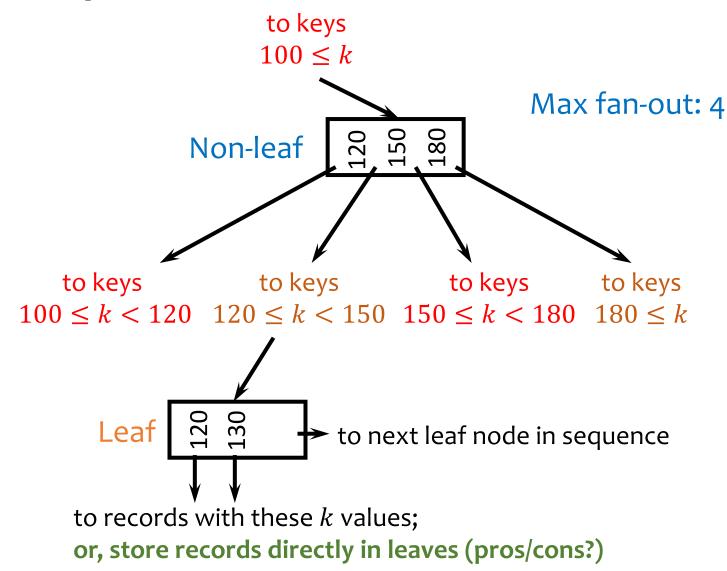
B⁺-tree: Closer Look

Max fan-out: 4

- A hierarchy of nodes with intervals
- Balanced (more or less): good performance guarantee
- Disk-based: one node per block; large fan-out



Sample B⁺-tree nodes



- Questions
- Why do we use B+-tree as database index instead of binary trees?



- Why do we use B⁺-tree as database index instead of B-trees?
 - What are the differences/pros/cons of B-trees vs. B⁺-tree as index?

B⁺-tree versus B-tree

- B-tree: why not store records (or record pointers) in non-leaf nodes?
 - These records can be accessed with fewer I/O's
- Problems?
 - Storing more data in a node decreases fan-out and increases h
 - Records in leaves require more I/O's to access
 - Vast majority of the records live in leaves!

39

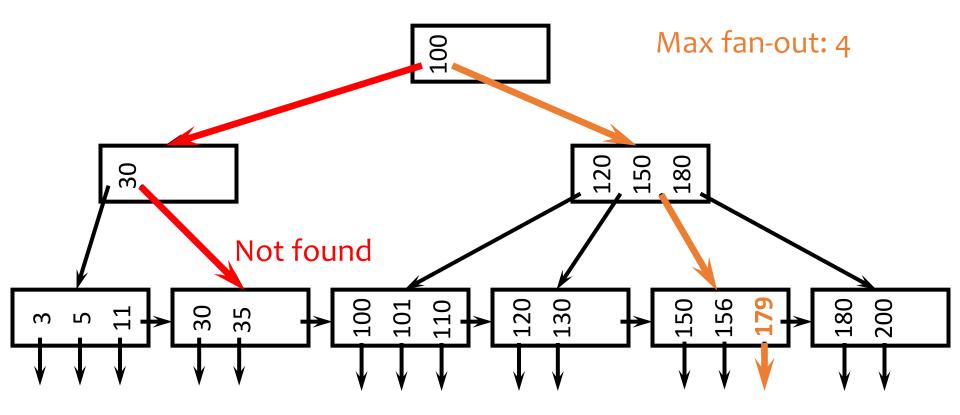
B⁺-tree balancing properties

- Height constraint: all leaves at the same lowest level
- Fan-out constraint: all nodes at least half full (except root)

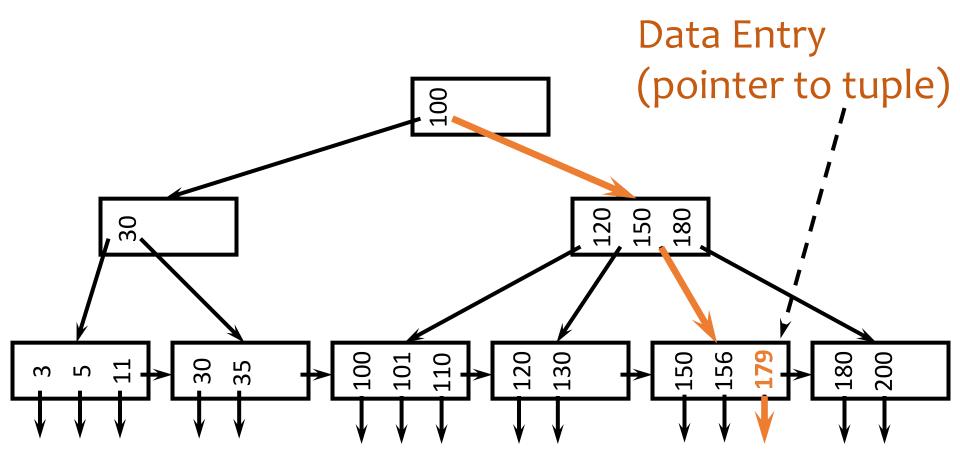
	Max #	Max #	Min #	Min #
	pointers	keys	active pointers	keys
Non-leaf	f f	f - 1	[<i>f</i> /2]	[f/2] - 1
Root	f	f - 1	2	1
Leaf	f	f - 1	$\lfloor f/2 \rfloor$	$\lfloor f/2 \rfloor$

Lookups

- SELECT * FROM *R* WHERE *k* = 179;
- SELECT * FROM *R* WHERE *k* = 32;

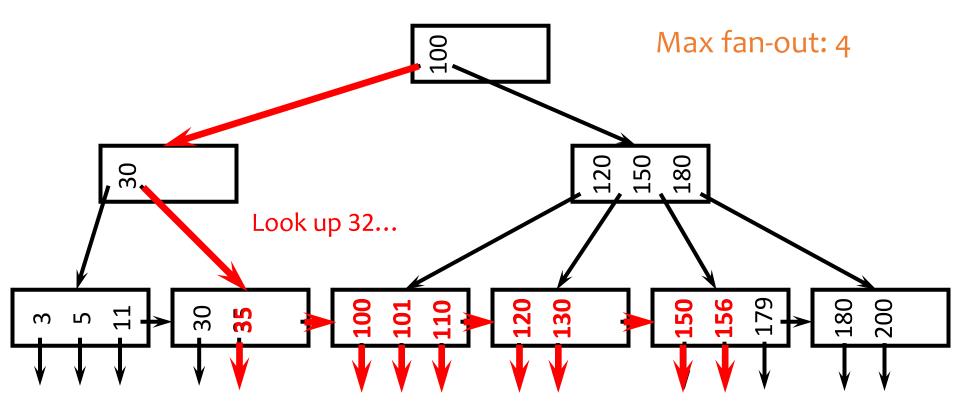


Search key and Data entry Search key Search key Select * FROM R WHERE k = 179; (value)



Range query

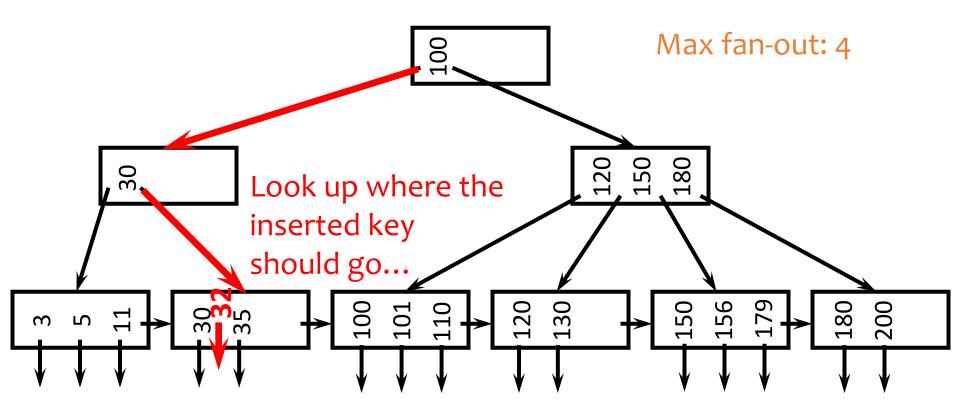
• SELECT * FROM *R* WHERE *k* > 32 AND *k* < 179;



And follow next-leaf pointers until you hit upper bound

Insertion

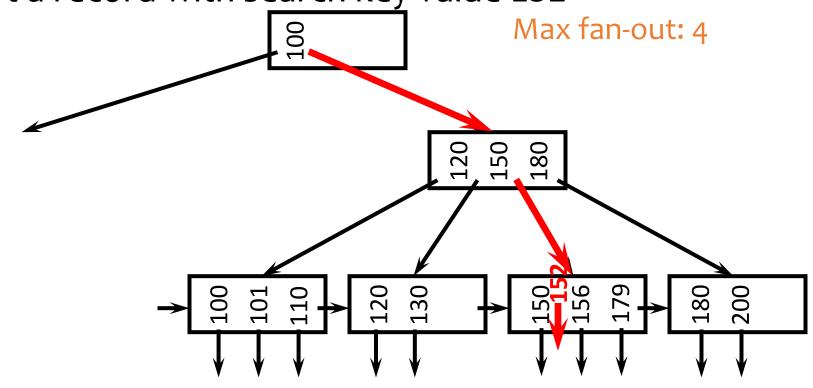
• Insert a record with search key value 32



And insert it right there

Another insertion example

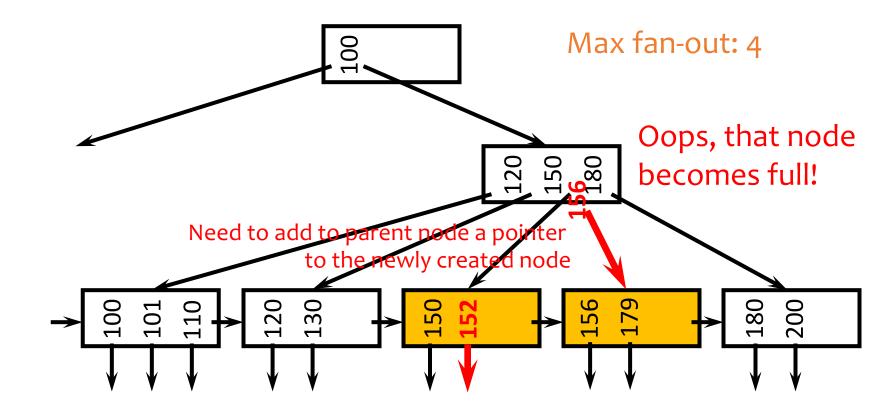
• Insert a record with search key value 152



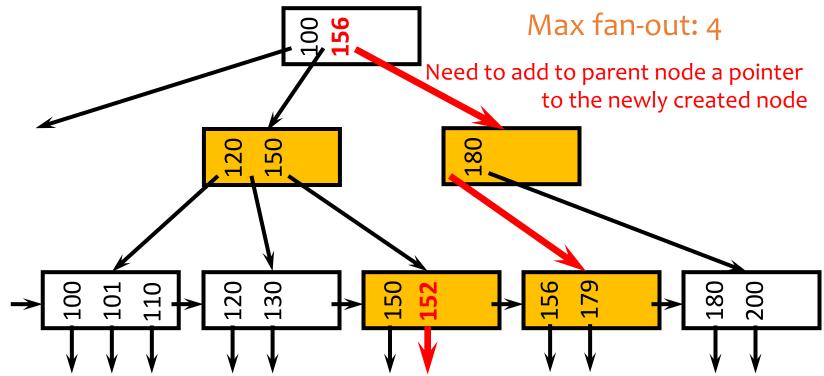
Oops, node is already full!

What are our options here?

Node splitting



More node splitting

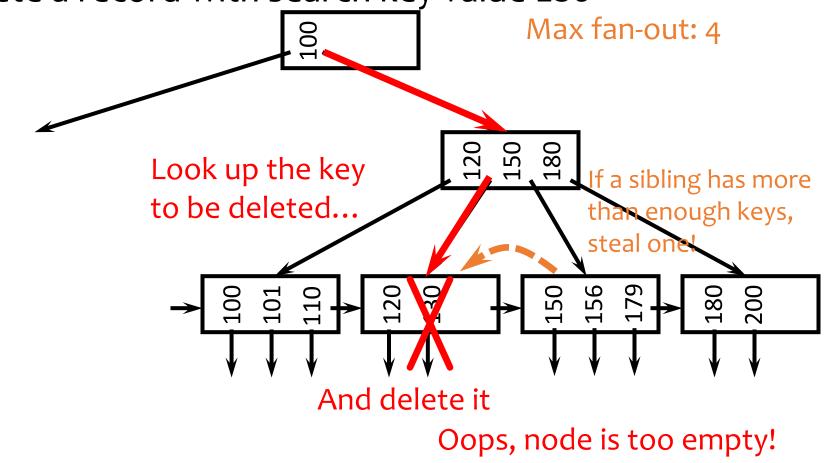


- In the worst case, node splitting can "propagate" all the way up to the root of the tree (not illustrated here)
 - Splitting the root introduces a new root of fan-out 2 and causes the tree to grow "up" by one level

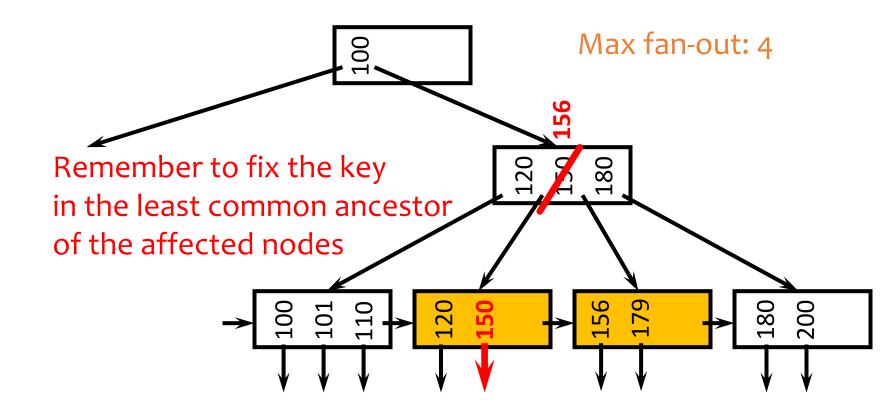
End of Lecture 10/6

Deletion

• Delete a record with search key value 130

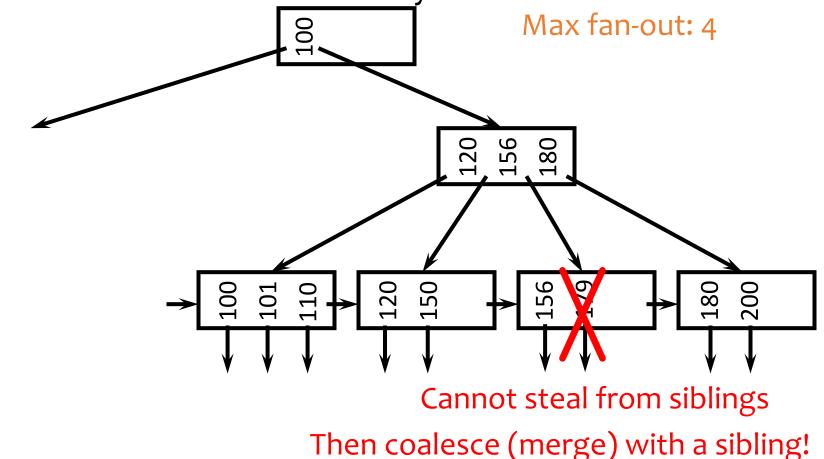


Stealing from a sibling

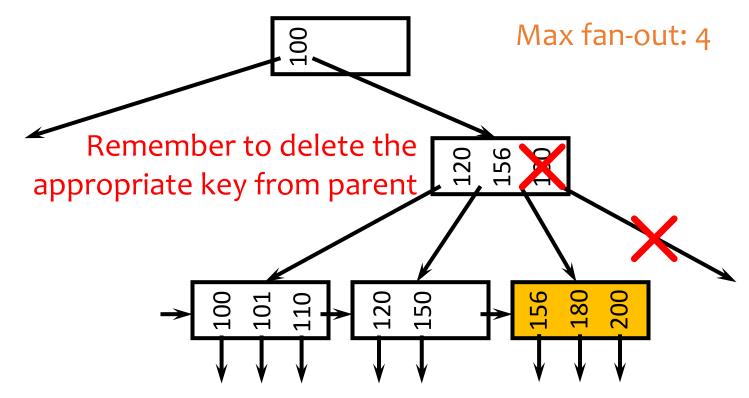


Another deletion example

• Delete a record with search key value 179



Coalescing



- Deletion can "propagate" all the way up to the root of the tree (not illustrated here)
 - When the root becomes empty, the tree "shrinks" by one level

Performance analysis

- How many I/O's are required for each operation?
 - *h*, the height of the tree (more or less)
 - Plus one or two to manipulate actual records
 - Plus O(h) for reorganization (rare if f is large)
 - Minus one if we cache the root in memory
- How big is *h*?
 - Roughly $\log_{fanout} N$, where N is the number of records
 - B⁺-tree properties guarantee that fan-out is least *f*/2 for all non-root nodes
 - Fan-out is typically large (in hundreds)—many keys and pointers can fit into one block
 - A 4-level B⁺-tree is enough for "typical" tables

B⁺-tree in practice

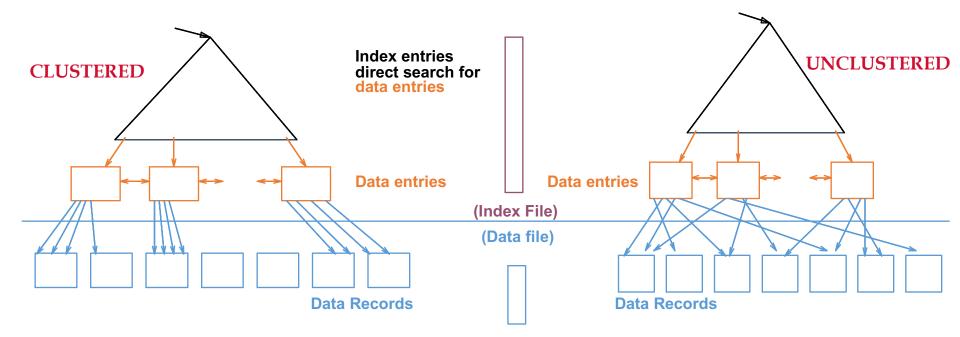
- Complex reorganization for deletion often is not implemented (e.g., Oracle)
 - Leave nodes less than half full and periodically reorganize
- Most commercial DBMS use B⁺-tree instead of hashing-based indexes because B⁺-tree handles range queries
 - A key difference between hash and tree indexes!

The Halloween Problem

- Story from the early days of System R...
 - UPDATE Payroll SET salary = salary * 1.1 WHERE salary <= 25000;
 - There is a B⁺-tree index on Payroll(salary)
 - All employees end up earning >= 25000 (why?)
- Solutions?
 - Scan index in reverse, or
 - Before update, scan index to create a "to-do" list, or
 - During update, maintain a "done" list, or
 - Tag every row with transaction/statement id

Clustered vs. Unclustered Index

- If order of data records in a file is the same as, or `close to', order of data entries in an index, then clustered, otherwise unclustered
- How does it affect # of page accesses? (in class)



Data is sorted on search key Data can be anywhere Clustered vs. Unclustered Index

- How does it affect # of page accesses?
 - Recall disk-memory diagram!
- SELECT * FROM USER WHERE age = 50
 - Assume 12 users with age = 50
 - Assume one data page can hold 4 User tuples
 - Suppose searching for a data entry requires 3 IOs in a B+-tree, which contain pointers to the data records (assume all matching pointers are in the same node of B+-tree)
 - What happens if the index is **unclustered**?
 - What happens if the index is clustered?

FYI – not covered in this class

Beyond ISAM, B-trees, and B⁺-trees

- Other tree-based indexes: R-trees and variants, GiST, etc.
- Hashing-based indexes: extensible hashing, linear hashing, etc.
- Text indexes: inverted-list index, suffix arrays, etc.
- Other tricks: bitmap index, bit-sliced index, etc.

Hash vs. Tree Index

- Hash indexes can only handle equality queries
 - SELECT * FROM R WHERE age = 5 (requires hash index on (age))
 - SELECT * FROM R, S WHERE R.A = S.A (requires hash index on R.A or S.A)
 - SELECT * FROM R WHERE age = 5 and name = 'Bart' (requires hash index on (age, name))
- (-) Cannot handle range queries or prefixes
 - SELECT * FROM R WHERE age >= 5
 - need to use tree indexes (more common)
 - Tree index on (age), or (age, name) works, but not (name, age) why?
- (+) Hash-indexes are more amenable to parallel processing
 - Will learn more in hash-based join
- Performance depends on how good the hash function is (whether the hash function distributes data uniformly and whether data has skew)

Trade-offs for Indexes

• Should we use as many indexes as possible?

Trade-offs for Indexes

- Should we use as many indexes as possible?
- Indexes can make
 - queries go faster
 - updates slower
- Require disk space, too

Index-Only Plans

 A number of queries can be answered without retrieving any tuples from one or more of the relations involved if a suitable index is available

SELECT E.dno, COUNT(*) FROM Emp E GROUP BY E.dno SELECT E.dno, MIN(E.sal) FROM Emp E GROUP BY E.dno

<E.dno,E.sal>

Tree index!

<*E.dno*>

<*E. age,E.sal*>

Tree index!

- If you have an index on E.dno in the above query, no need to access data
- For index-only strategies, clustering is not important

SELECT AVG(E.sal) FROM Emp E WHERE E.age=25 AND E.sal BETWEEN 3000 AND 5000